

## Savitch's Theorem

### Theorem 25 (Savitch (1970))

REACHABILITY  $\in$  SPACE( $\log^2 n$ ).

- Let  $G(V, E)$  be a graph with  $n$  nodes.
- For  $i \geq 0$ , let

PATH( $x, y, i$ )

mean there is a path from node  $x$  to node  $y$  of length at most  $2^i$ .

- There is a path from  $x$  to  $y$  if and only if

PATH( $x, y, \lceil \log n \rceil$ )

holds.

## The Proof (continued)

- For  $i > 0$ ,  $\text{PATH}(x, y, i)$  if and only if there exists a  $z$  such that  $\text{PATH}(x, z, i - 1)$  and  $\text{PATH}(z, y, i - 1)$ .
- For  $\text{PATH}(x, y, 0)$ , check the input graph or if  $x = y$ .
- Compute  $\text{PATH}(x, y, \lceil \log n \rceil)$  with a depth-first search on a graph with nodes  $(x, y, z, i)$ s (see next page).<sup>a</sup>
- Like stacks in recursive calls, we keep only the current path of  $(x, y, i)$ s.
- The space requirement is proportional to the depth of the tree ( $\lceil \log n \rceil$ ) times the size of the items stored at each node.

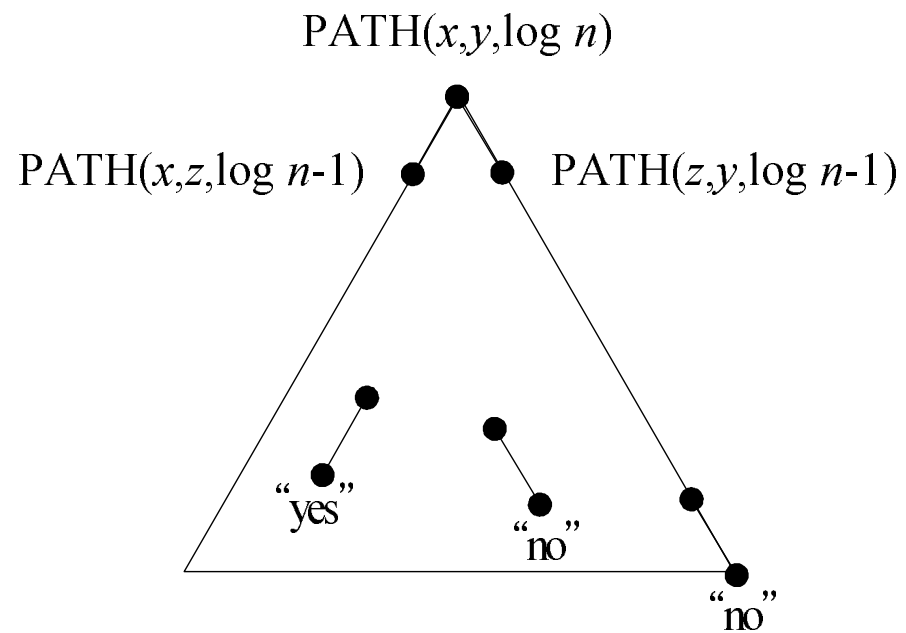
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<sup>a</sup>Contributed by Mr. Chuan-Yao Tan on October 11, 2011.

## The Proof (continued): Algorithm for $\text{PATH}(x, y, i)$

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1: if  $i = 0$  then  
2:   if  $x = y$  or  $(x, y) \in E$  then  
3:     return true;  
4:   else  
5:     return false;  
6:   end if  
7: else  
8:   for  $z = 1, 2, \dots, n$  do  
9:     if  $\text{PATH}(x, z, i - 1)$  and  $\text{PATH}(z, y, i - 1)$  then  
10:      return true;  
11:    end if  
12:  end for  
13:  return false;  
14: end if
```

## The Proof (continued)



## The Proof (concluded)

- Depth is  $\lceil \log n \rceil$ , and each node  $(x, y, z, i)$  needs space  $O(\log n)$ .
- The total space is  $O(\log^2 n)$ .

## The Relation between Nondeterministic Space and Deterministic Space Only Quadratic

**Corollary 26** *Let  $f(n) \geq \log n$  be proper. Then*

$$\text{NSPACE}(f(n)) \subseteq \text{SPACE}(f^2(n)).$$

- Apply Savitch's proof to the configuration graph of the NTM on the input.
- From p. 242, the configuration graph has  $O(c^{f(n)})$  nodes; hence each node takes space  $O(f(n))$ .
- But if we construct *explicitly* the whole graph before applying Savitch's theorem, we get  $O(c^{f(n)})$  space!

## The Proof (continued)

- The way out is *not* to generate the graph at all.
- Instead, keep the graph implicit.
- In fact, we check node connectedness only when  $i = 0$  on p. 250, by examining the input string  $G$ .
- There, given configurations  $x$  and  $y$ , we go over the Turing machine's program to determine if there is an instruction that can turn  $x$  into  $y$  in one step.<sup>a</sup>

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<sup>a</sup>Thanks to a lively class discussion on October 15, 2003.

## The Proof (concluded)

- The  $z$  variable in the algorithm on p. 250 simply runs through all possible valid configurations.
  - Let  $z = 0, 1, \dots, O(c^{f(n)})$ .
  - Make sure  $z$  is a valid configuration before using it in the recursive calls.<sup>a</sup>
- Each  $z$  has length  $O(f(n))$  by Eq. (2) on p. 242.
- So each node needs space  $O(f(n))$ .
- The depth of the recursive call on p. 250 is  $O(\log c^{f(n)})$ , which is  $O(f(n))$ .
- The total space is therefore  $O(f^2(n))$ .

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<sup>a</sup>Thanks to a lively class discussion on October 13, 2004.



## Implications of Savitch's Theorem

- $PSPACE = NPSPACE$ .
- Nondeterminism is less powerful with respect to space.
- Nondeterminism may be very powerful with respect to time as it is not known if  $P = NP$ .

## Nondeterministic Space Is Closed under Complement

- Closure under complement is trivially true for deterministic complexity classes (p. 227).
- It is known that<sup>a</sup>

$$\text{coNSPACE}(f(n)) = \text{NSPACE}(f(n)). \quad (3)$$

- So

$$\begin{aligned} \text{coNL} &= \text{NL}, \\ \text{coNPSPACE} &= \text{NPSPACE}. \end{aligned}$$

- But it is not known whether  $\text{coNP} = \text{NP}$ .

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<sup>a</sup>Szelepcényi (1987) and Immerman (1988).

# *Reductions and Completeness*

It is unworthy of excellent men  
to lose hours like slaves in the labor of  
computation.  
— Gottfried Wilhelm von Leibniz (1646–1716)

## Degrees of Difficulty

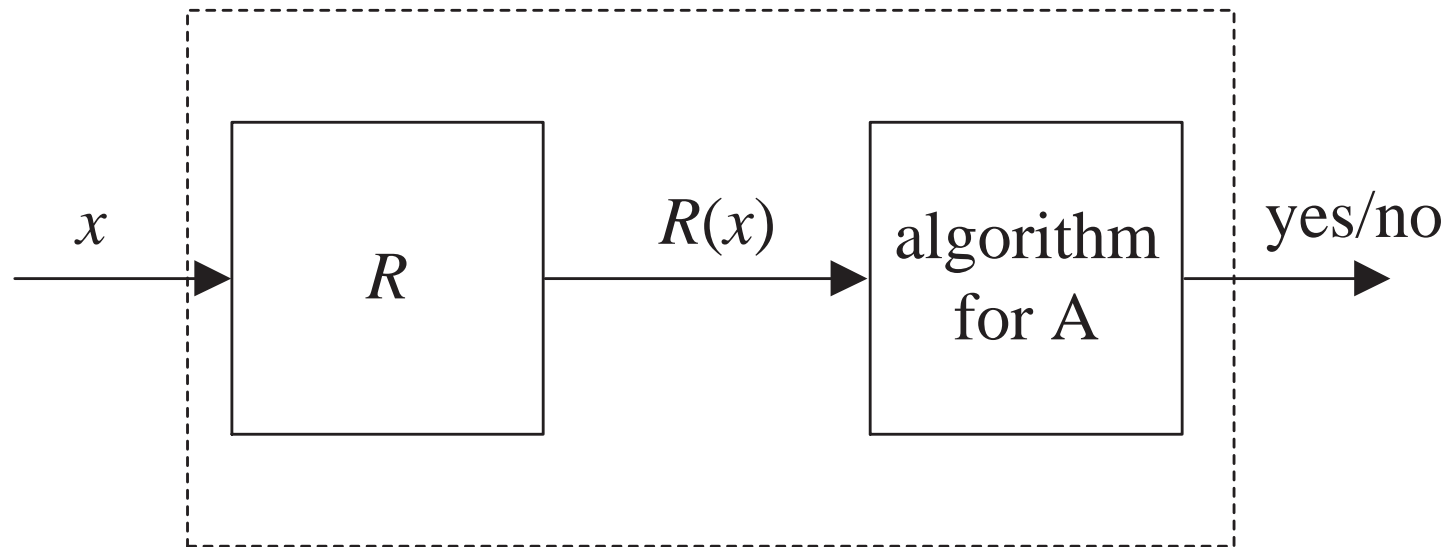
- When is a problem more difficult than another?
- **B reduces to A** if there is a transformation  $R$  which for every input  $x$  of B yields an input  $R(x)$  of A.<sup>a</sup>
  - The answer to  $x$  for B is the same as the answer to  $R(x)$  for A.
  - $R$  is easy to compute.
- We say problem A is at least as hard as<sup>b</sup> problem B if B reduces to A.

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<sup>a</sup>See also p. 164.

<sup>b</sup>Or simply “harder than” for brevity.

## Reduction



Solving problem B by calling the algorithm for problem A *once* and *without* further processing its answer.

## Degrees of Difficulty (concluded)

- This makes intuitive sense: If  $A$  is able to solve your problem  $B$  after only a little bit of work of  $R$ , then  $A$  must be at least as hard.
  - If  $A$  is easy to solve, it combined with  $R$  (which is also easy) would make  $B$  easy to solve, too.<sup>a</sup>
  - So if  $B$  is hard to solve,  $A$  must be hard (if not harder), too!

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<sup>a</sup>Thanks to a lively class discussion on October 13, 2009.

## Comments<sup>a</sup>

- Suppose  $B$  reduces to  $A$  via a transformation  $R$ .
- The input  $x$  is an instance of  $B$ .
- The output  $R(x)$  is an instance of  $A$ .
- $R(x)$  may not span all possible instances of  $A$ .<sup>b</sup>
  - Some instances of  $A$  may never appear in the range of  $R$ .
- But  $x$  must be a general instance for  $B$ .

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<sup>a</sup>Contributed by Mr. Ming-Feng Tsai (D92922003) on October 29, 2003.

<sup>b</sup> $R(x)$  may not be onto; Mr. Alexandr Simak (D98922040) on October 13, 2009.



## Is “Reduction” a Confusing Choice of Word?<sup>a</sup>

- If B reduces to A, doesn't that intuitively make A smaller and simpler?
  - Sometimes, we say, “B can be reduced to A.”
- But our definition means just the opposite.
- Our definition says in this case B is a special case of A.
- Hence A is harder.

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<sup>a</sup>Moore and Mertens (2011).

## Reduction between Languages

- Language  $L_1$  is **reducible to**  $L_2$  if there is a function  $R$  computable by a deterministic TM in space  $O(\log n)$ .
- Furthermore, for all inputs  $x$ ,  $x \in L_1$  if and only if  $R(x) \in L_2$ .
- $R$  is said to be a **(Karp) reduction** from  $L_1$  to  $L_2$ .

## Reduction between Languages (concluded)

- Note that by Theorem 24 (p. 239),  $R$  runs in polynomial time.
  - In most cases, a polynomial-time  $R$  suffices for proofs.<sup>a</sup>
- Suppose  $R$  is a reduction from  $L_1$  to  $L_2$ .
- Then solving “ $R(x) \in L_2?$ ” is an algorithm for solving “ $x \in L_1?$ ”<sup>b</sup>

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<sup>a</sup>In fact, unless stated otherwise, we will only require that the reduction  $R$  run in polynomial time.

<sup>b</sup>Of course, it may not be an optimal one.

## A Paradox?

- Degree of difficulty is not defined in terms of *absolute* complexity.
- So a language  $B \in \text{TIME}(n^{99})$  may be “easier” than a language  $A \in \text{TIME}(n^3)$ .
  - Again, this happens when B is reducible to A.
- But isn't this a contradiction if the best algorithm for B requires  $n^{99}$  steps?
- That is, how can a problem *requiring*  $n^{99}$  steps be reducible to a problem solvable in  $n^3$  steps?

## Paradox Resolved

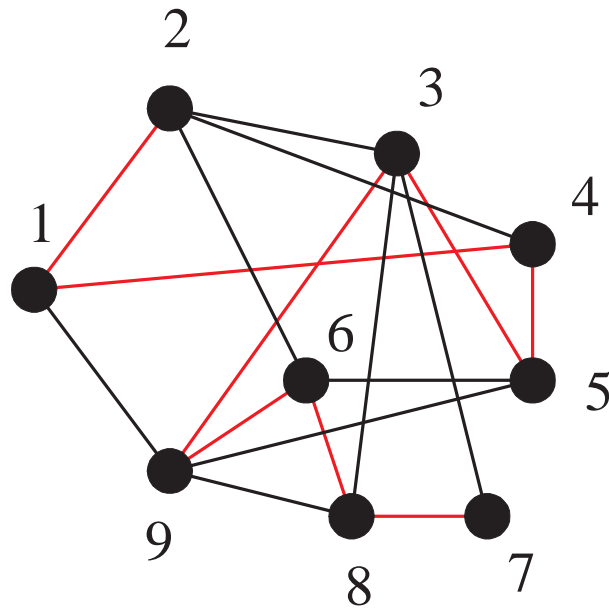
- The so-called contradiction does not hold.
- Suppose we solve the problem “ $x \in B?$ ” via “ $R(x) \in A?$ ”
- We must consider the time spent by  $R(x)$  and its length  $|R(x)|$ :
  - Because  $R(x)$  (not  $x$ ) is solved by  $A$ .

## HAMILTONIAN PATH

- A **Hamiltonian path** of a graph is a path that visits every node of the graph exactly once.
- Suppose graph  $G$  has  $n$  nodes:  $1, 2, \dots, n$ .
- A Hamiltonian path can be expressed as a permutation  $\pi$  of  $\{1, 2, \dots, n\}$  such that
  - $\pi(i) = j$  means the  $i$ th position is occupied by node  $j$ .
  - $(\pi(i), \pi(i + 1)) \in G$  for  $i = 1, 2, \dots, n - 1$ .
- HAMILTONIAN PATH asks if a graph has a Hamiltonian path.

## Reduction of HAMILTONIAN PATH to SAT

- Given a graph  $G$ , we shall construct a CNF  $R(G)$  such that  $R(G)$  is satisfiable iff  $G$  has a Hamiltonian path.
- $R(G)$  has  $n^2$  boolean variables  $x_{ij}$ ,  $1 \leq i, j \leq n$ .
- $x_{ij}$  means  
“the  $i$ th position in the Hamiltonian path is occupied by node  $j$ .”
- Our reduction will produce clauses.



$$\begin{aligned}
 &x_{12} = x_{21} = x_{34} = x_{45} = x_{53} = x_{69} = x_{76} = x_{88} = x_{97} = 1; \\
 &\pi(1) = 2, \pi(2) = 1, \pi(3) = 4, \pi(4) = 5, \pi(5) = 3, \pi(6) = \\
 &9, \pi(7) = 6, \pi(8) = 8, \pi(9) = 7.
 \end{aligned}$$



## The Clauses of $R(G)$ and Their Intended Meanings

1. Each node  $j$  must appear in the path.
  - $x_{1j} \vee x_{2j} \vee \cdots \vee x_{nj}$  for each  $j$ .
2. No node  $j$  appears twice in the path.
  - $\neg x_{ij} \vee \neg x_{kj} (\equiv \neg(x_{ij} \wedge x_{kj}))$  for all  $i, j, k$  with  $i \neq k$ .
3. Every position  $i$  on the path must be occupied.
  - $x_{i1} \vee x_{i2} \vee \cdots \vee x_{in}$  for each  $i$ .
4. No two nodes  $j$  and  $k$  occupy the same position in the path.
  - $\neg x_{ij} \vee \neg x_{ik} (\equiv \neg(x_{ij} \wedge x_{ik}))$  for all  $i, j, k$  with  $j \neq k$ .
5. Nonadjacent nodes  $i$  and  $j$  cannot be adjacent in the path.
  - $\neg x_{ki} \vee \neg x_{k+1,j}$  for all  $(i, j) \notin G$  and  $k = 1, 2, \dots, n - 1$ .

## The Proof

- $R(G)$  contains  $O(n^3)$  clauses.
- $R(G)$  can be computed efficiently (simple exercise).
- Suppose  $T \models R(G)$ .
- From the 1st and 2nd types of clauses, for each node  $j$  there is a unique position  $i$  such that  $T \models x_{ij}$ .
- From the 3rd and 4th types of clauses, for each position  $i$  there is a unique node  $j$  such that  $T \models x_{ij}$ .
- So there is a permutation  $\pi$  of the nodes such that  $\pi(i) = j$  if and only if  $T \models x_{ij}$ .

## The Proof (concluded)

- The 5th type of clauses furthermore guarantee that  $(\pi(1), \pi(2), \dots, \pi(n))$  is a Hamiltonian path.
- Conversely, suppose  $G$  has a Hamiltonian path

$$(\pi(1), \pi(2), \dots, \pi(n)),$$

where  $\pi$  is a permutation.

- Clearly, the truth assignment

$$T(x_{ij}) = \mathbf{true} \text{ if and only if } \pi(i) = j$$

satisfies all clauses of  $R(G)$ .

## A Comment<sup>a</sup>

- An answer to “Is  $R(G)$  satisfiable?” does answer “Is  $G$  Hamiltonian?”
- But a positive answer does not give a Hamiltonian path for  $G$ .
  - Providing a witness is not a requirement of reduction.
- A positive answer to “Is  $R(G)$  satisfiable?” *plus a satisfying truth assignment* does provide us with a Hamiltonian path for  $G$ .

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<sup>a</sup>Contributed by Ms. Amy Liu (J94922016) on May 29, 2006.

## Reduction of REACHABILITY to CIRCUIT VALUE

- Note that both problems are in P.
- Given a graph  $G = (V, E)$ , we shall construct a *variable-free* circuit  $R(G)$ .
- The output of  $R(G)$  is true if and only if there is a path from node 1 to node  $n$  in  $G$ .
- Idea: the Floyd-Warshall algorithm.

## The Gates

- The gates are
  - $g_{ijk}$  with  $1 \leq i, j \leq n$  and  $0 \leq k \leq n$ .
  - $h_{ijk}$  with  $1 \leq i, j, k \leq n$ .
- $g_{ijk}$ : There is a path from node  $i$  to node  $j$  without passing through a node bigger than  $k$ .
- $h_{ijk}$ : There is a path from node  $i$  to node  $j$  passing through  $k$  but not any node bigger than  $k$ .
- Input gate  $g_{ij0} = \text{true}$  if and only if  $i = j$  or  $(i, j) \in E$ .

## The Construction

- $h_{ijk}$  is an AND gate with predecessors  $g_{i,k,k-1}$  and  $g_{k,j,k-1}$ , where  $k = 1, 2, \dots, n$ .
- $g_{ijk}$  is an OR gate with predecessors  $g_{i,j,k-1}$  and  $h_{i,j,k}$ , where  $k = 1, 2, \dots, n$ .
- $g_{1nn}$  is the output gate.
- Interestingly,  $R(G)$  uses no  $\neg$  gates.
  - It is a **monotone circuit**.

## Reduction of CIRCUIT SAT to SAT

- Given a circuit  $C$ , we will construct a boolean expression  $R(C)$  such that  $R(C)$  is satisfiable iff  $C$  is.
  - $R(C)$  will turn out to be a CNF.
  - $R(C)$  is basically a depth-2 circuit; furthermore, each gate has out-degree 1.
- The variables of  $R(C)$  are those of  $C$  plus  $g$  for each gate  $g$  of  $C$ .
  - The  $g$ 's propagate the truth values for the CNF.
- Each gate of  $C$  will be turned into equivalent clauses.
- Recall that clauses are  $\wedge$ ed together by definition.



## The Clauses of $R(C)$

**$g$  is a variable gate  $x$ :** Add clauses  $(\neg g \vee x)$  and  $(g \vee \neg x)$ .

- Meaning:  $g \Leftrightarrow x$ .

**$g$  is a true gate:** Add clause  $(g)$ .

- Meaning:  $g$  must be true to make  $R(C)$  true.

**$g$  is a false gate:** Add clause  $(\neg g)$ .

- Meaning:  $g$  must be false to make  $R(C)$  true.

**$g$  is a  $\neg$  gate with predecessor gate  $h$ :** Add clauses  $(\neg g \vee \neg h)$  and  $(g \vee h)$ .

- Meaning:  $g \Leftrightarrow \neg h$ .

## The Clauses of $R(C)$ (concluded)

**$g$  is a  $\vee$  gate with predecessor gates  $h$  and  $h'$ :** Add clauses  $(\neg h \vee g)$ ,  $(\neg h' \vee g)$ , and  $(h \vee h' \vee \neg g)$ .

- Meaning:  $g \Leftrightarrow (h \vee h')$ .

**$g$  is a  $\wedge$  gate with predecessor gates  $h$  and  $h'$ :** Add clauses  $(\neg g \vee h)$ ,  $(\neg g \vee h')$ , and  $(\neg h \vee \neg h' \vee g)$ .

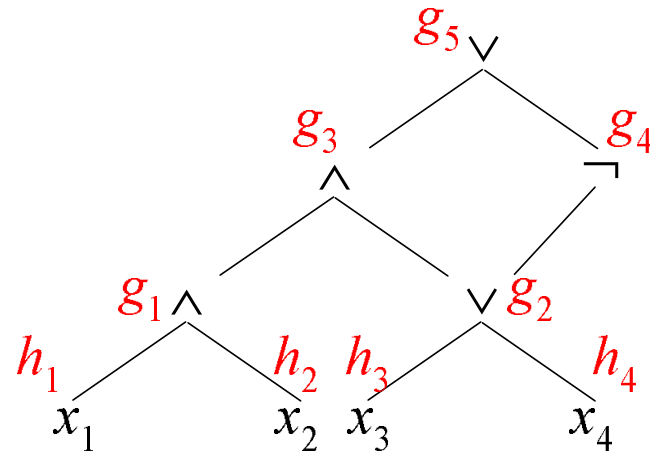
- Meaning:  $g \Leftrightarrow (h \wedge h')$ .

**$g$  is the output gate:** Add clause  $(g)$ .

- Meaning:  $g$  must be true to make  $R(C)$  true.

Note: If gate  $g$  feeds gates  $h_1, h_2, \dots$ , then variable  $g$  appears in the clauses for  $h_1, h_2, \dots$  in  $R(C)$ .

## An Example



$$\begin{aligned}
 & (h_1 \Leftrightarrow x_1) \wedge (h_2 \Leftrightarrow x_2) \wedge (h_3 \Leftrightarrow x_3) \wedge (h_4 \Leftrightarrow x_4) \\
 \wedge & [g_1 \Leftrightarrow (h_1 \wedge h_2)] \wedge [g_2 \Leftrightarrow (h_3 \vee h_4)] \\
 \wedge & [g_3 \Leftrightarrow (g_1 \wedge g_2)] \wedge (g_4 \Leftrightarrow \neg g_2) \\
 \wedge & [g_5 \Leftrightarrow (g_3 \vee g_4)] \wedge g_5.
 \end{aligned}$$

## An Example (concluded)

- In general, the result is a CNF.
- The CNF has size proportional to the circuit's number of gates.
- The CNF adds new variables to the circuit's original input variables.
- Had we used the idea on p. 209 for the reduction, the resulting formula may have an exponential length because of the copying.<sup>a</sup>

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<sup>a</sup>Contributed by Mr. Ching-Hua Yu (D00921025) on October 16, 2012.

## Composition of Reductions

**Proposition 27** *If  $R_{12}$  is a reduction from  $L_1$  to  $L_2$  and  $R_{23}$  is a reduction from  $L_2$  to  $L_3$ , then the composition  $R_{12} \circ R_{23}$  is a reduction from  $L_1$  to  $L_3$ .*

- So reducibility is transitive.

## Completeness<sup>a</sup>

- As reducibility is transitive, problems can be ordered with respect to their difficulty.
- Is there a *maximal* element (the *hardest* problem)?
- It is not obvious that there should be a maximal element.
  - Many infinite structures (such as integers and real numbers) do not have maximal elements.
- Hence it may surprise you that most of the complexity classes that we have seen so far have maximal elements.

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<sup>a</sup>Cook (1971) and Levin (1973).

## Completeness (concluded)

- Let  $\mathcal{C}$  be a complexity class and  $L \in \mathcal{C}$ .
- $L$  is  **$\mathcal{C}$ -complete** if every  $L' \in \mathcal{C}$  can be reduced to  $L$ .
  - Most complexity classes we have seen so far have complete problems!
- Complete problems capture the difficulty of a class because they are the hardest problems in the class.

## Hardness

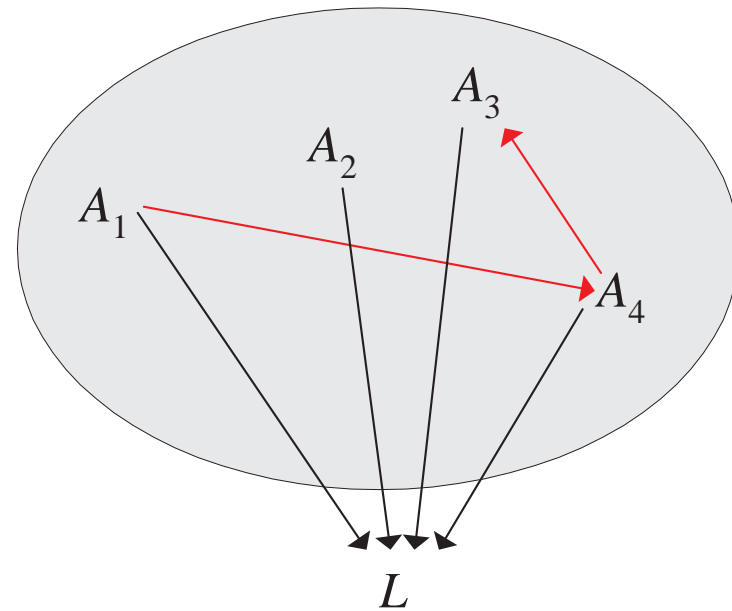
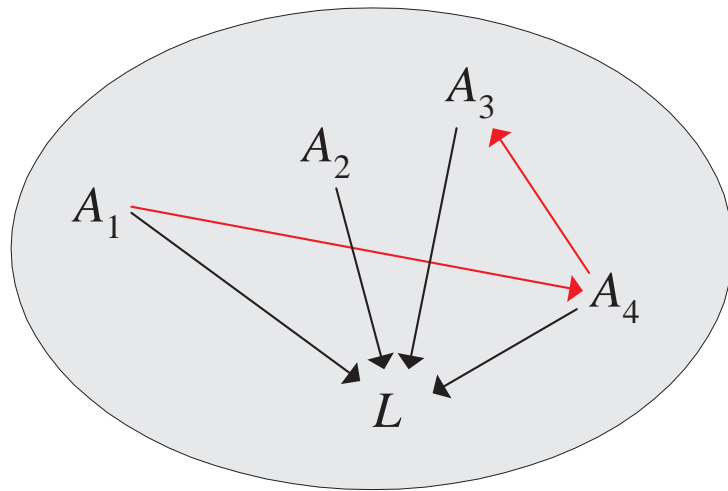
- Let  $\mathcal{C}$  be a complexity class.
- $L$  is  **$\mathcal{C}$ -hard** if every  $L' \in \mathcal{C}$  can be reduced to  $L$ .
- It is not required that  $L \in \mathcal{C}$ .
- If  $L$  is  $\mathcal{C}$ -hard, then by definition, every  $\mathcal{C}$ -complete problem can be reduced to  $L$ .<sup>a</sup>

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<sup>a</sup>Contributed by Mr. Ming-Feng Tsai (D92922003) on October 15, 2003.



## Illustration of Completeness and Hardness



## Closedness under Reductions

- A class  $\mathcal{C}$  is **closed under reductions** if whenever  $L$  is reducible to  $L'$  and  $L' \in \mathcal{C}$ , then  $L \in \mathcal{C}$ .
- It is easy to show that P, NP, coNP, L, NL, PSPACE, and EXP are all closed under reductions.

## Complete Problems and Complexity Classes

**Proposition 28** *Let  $\mathcal{C}'$  and  $\mathcal{C}$  be two complexity classes such that  $\mathcal{C}' \subseteq \mathcal{C}$ . Assume  $\mathcal{C}'$  is closed under reductions and  $L$  is  $\mathcal{C}$ -complete. Then  $\mathcal{C} = \mathcal{C}'$  if and only if  $L \in \mathcal{C}'$ .*

- Suppose  $L \in \mathcal{C}'$  first.
- Every language  $A \in \mathcal{C}$  reduces to  $L \in \mathcal{C}'$ .
- Because  $\mathcal{C}'$  is closed under reductions,  $A \in \mathcal{C}'$ .
- Hence  $\mathcal{C} \subseteq \mathcal{C}'$ .
- As  $\mathcal{C}' \subseteq \mathcal{C}$ , we conclude that  $\mathcal{C} = \mathcal{C}'$ .

## The Proof (concluded)

- On the other hand, suppose  $\mathcal{C} = \mathcal{C}'$ .
- As  $L$  is  $\mathcal{C}$ -complete,  $L \in \mathcal{C}$ .
- Thus, trivially,  $L \in \mathcal{C}'$ .

## Two Important Corollaries

Proposition 28 implies the following.

**Corollary 29**  *$P = NP$  if and only if an NP-complete problem is in  $P$ .*

**Corollary 30**  *$L = P$  if and only if a P-complete problem is in  $L$ .*

## Complete Problems and Complexity Classes

**Proposition 31** *Let  $\mathcal{C}'$  and  $\mathcal{C}$  be two complexity classes closed under reductions. If  $L$  is complete for both  $\mathcal{C}$  and  $\mathcal{C}'$ , then  $\mathcal{C} = \mathcal{C}'$ .*

- All languages  $\mathcal{L} \in \mathcal{C}$  reduce to  $L \in \mathcal{C}$  and  $L \in \mathcal{C}'$ .
- Since  $\mathcal{C}'$  is closed under reductions,  $\mathcal{L} \in \mathcal{C}'$ .
- Hence  $\mathcal{C} \subseteq \mathcal{C}'$ .
- The proof for  $\mathcal{C}' \subseteq \mathcal{C}$  is symmetric.

## Table of Computation

- Let  $M = (K, \Sigma, \delta, s)$  be a single-string polynomial-time deterministic TM deciding  $L$ .
- Its computation on input  $x$  can be thought of as a  $|x|^k \times |x|^k$  table, where  $|x|^k$  is the time bound.
  - It is essentially a sequence of configurations.
- Rows correspond to time steps 0 to  $|x|^k - 1$ .
- Columns are positions in the string of  $M$ .
- The  $(i, j)$ th table entry represents the contents of position  $j$  of the string *after*  $i$  steps of computation.

## Some Conventions To Simplify the Table

- $M$  halts after at most  $|x|^k - 2$  steps.
- Assume a large enough  $k$  to make it true for  $|x| \geq 2$ .
- Pad the table with  $\sqcup$ s so that each row has length  $|x|^k$ .
  - The computation will never reach the right end of the table for lack of time.
- If the cursor scans the  $j$ th position at time  $i$  when  $M$  is at state  $q$  and the symbol is  $\sigma$ , then the  $(i, j)$ th entry is a *new* symbol  $\sigma_q$ .



## Some Conventions To Simplify the Table (continued)

- If  $q$  is “yes” or “no,” simply use “yes” or “no” instead of  $\sigma_q$ .
- Modify  $M$  so that the cursor starts not at  $\triangleright$  but at the first symbol of the input.
- The cursor never visits the leftmost  $\triangleright$  by telescoping two moves of  $M$  each time the cursor is about to move to the leftmost  $\triangleright$ .
- So the first symbol in every row is a  $\triangleright$  and not a  $\triangleright_q$ .

## Some Conventions To Simplify the Table (concluded)

- Suppose  $M$  has halted before its time bound of  $|x|^k$ , so that “yes” or “no” appears at a row before the last.
- Then all subsequent rows will be identical to that row.
- $M$  accepts  $x$  if and only if the  $(|x|^k - 1, j)$ th entry is “yes” for some position  $j$ .

## Comments

- Each row is essentially a configuration.
- If the input  $x = 010001$ , then the first row is

$$\begin{array}{c} |x|^k \\ \hline \triangleright 0_s 10001 \square \square \cdots \square \end{array}$$

- A typical row may look like

$$\begin{array}{c} |x|^k \\ \hline \triangleright 10100_q 01110100 \square \square \cdots \square \end{array}$$

## Comments (concluded)

- The last rows must look like

$$\overbrace{\triangleright \dots \text{“yes”} \dots \square}^{|x|^k} \quad \text{or} \quad \overbrace{\triangleright \dots \text{“no”} \dots \square}^{|x|^k}$$

- Three out of the table's 4 borders are known:

$$\begin{array}{cccccc} \triangleright & \mathbf{a} & \mathbf{b} & \mathbf{c} & \mathbf{d} & \mathbf{e} & \mathbf{f} & \square \\ \triangleright & & & & & & & \square \\ \triangleright & & & & & & & \square \\ \triangleright & & & & & & & \square \\ \triangleright & & & & & & & \square \\ & & & & \vdots & & & \square \end{array}$$